1.

Practitioner's Docket No. 740789-052110

09/980770

CHAPTER II

TO THE UNITED STATES ELECTED OFFICE (EO/US) (ENTRY INTO U.S. NATIONAL PHASE UNDER CHAPTER II)

INTERNATIONAL APPLICATION NO.

INTERNATIONAL FILING DATE

PRIORITY DATE CLAIMED

PCT/GB00/01675

02 May 2000 (02.05.00) i

01 May 1999 (01.05.99)

TITLE OF INVENTION

METHOD OF ANALYSIS OF MEDICAL SIGNALS

APPLICANTS

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Box PCT

Assistant Commissioner for Patents

Washington D.C. 20231

ATTENTION: EO/US

35 U.S.C. 371:

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Date: 01 November 2001

- Applicant herewith submits to the United States Elected Office (EO/US) the following items under
- This express request to immediately begin national examination procedures (35 Xa. U.S.C. Section 371(f)).
- The U.S. National Fee (35 U.S.C. Section 371(c)(1)) and other fees (37 C.F.R. [X]b. Section 1.492) as indicated below:

In re application of: Application No.:

ADDISON, P.S., et al. Not Yet Assigned

Group: Examiner: Not yet assigned Not yet assigned

Аррисанс

(National Phase Entry of PCT/GB00/01675) Herewith

Filed: For:

METHOD OF ANALYSIS OF MEDICAL SIGNALS

2.Fees

CLAIMS FEE	(1) FOR	(2) NUMBER FILED	(3) NUMBER EXTRA	(4) RATE		CALCULA- TIONS
[]*	TOTAL CLAIMS	- 20 =	0	x \$18.00 =	\$	0
	INDEPENDENT CLAIMS	-3=	0	x \$84.00 =	\$	0
	MULTIPLE DEPE	NDENT CLAIM(S) (if	applicable) + \$280.0	00	\$	0
BASIC FEE**	MULTIPLE DEPENDENT CLAIM(S) (if applicable) + \$280.00 \$ [] U.S. PTO WAS INTERNATIONAL PRELIMINARY EXAMINATION AUTHORITY Where an International preliminary examination fee as set forth in Section 1.482 has been paid on the international application to the U.S. PTO: [] and the international preliminary examination report states that the criteria of novelty, inventive step (non-obviousness) and industrial activity, as defined in PCT Article 33(2) to (4) have been satisfied for all the claims presented in the application entering the national stage (37 C.F.R. Section 1.492(a)(4)) [] and the above requirements are not met (37 C.F.R. Section 1.492(a)(1))					
		Office (37 C.F.R. Sect	ion 1.492(a)(5))	\$890.00	\$8	90.00
			Total	of above Calculations	=\$	890.00
SMALL ENTITY	Reduction by 1/2 for entity status.	or filing by small entity,	if applicable. Applicable.	cant asserts small	- \$	445.00
				Subtotal	\$	445.00
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i. [X] A check in the amount of \$\\$445.00\$ to cover the above fees is enclosed.
 ii. [X] Please charge Account No. 50-0850 - 37 C.F.R. Section 1.492(b), (c) and (d)

(presentation of extra claims)

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_	ation No.:	Not Yet Assigned Examiner: Not yet assigned
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3.	[X]	A copy of the International application as filed (35 U.S.C. Section 371(c)(2)):
	a.	[] is transmitted herewith.
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7.	[]	A copy [] []	is transn	nitted herewith	•	(PCT/IPEA/409) iled with the United States Receiving
8.	[] a. b. c.	Annex([] []	is/are tra *Applica Applica is/are no Office.	ansmitted herevants request ention if such has	try of the annex not been done.	to the IPER to the International
9.	[] a. b. c.	A trans [] [] []	is transi	nitted herewith	l.	al preliminary examination report e English language.
10.	[X] a.	An oat U.S.C.	115		nventor (35 U.S ted by applicant	.C. Section 371(c)(4)) complying with 35 t on
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L 3	An Information Disclosure Statement under 37 C.F.R. Sections 1.97 and 1.98: [] is transmitted herewith. Also transmitted herewith is/are: [] Form PTO-1449 (PTO/SB/08A and 08B). [] Copies of citations listed.	
b. с.	will be transmitted within THREE MONTHS of the date of submission of requirements under 35 U.S.C. Sections 371(c). was previously submitted by applicant on	
13. []	An assignment document is transmitted herewith for recording.	
A separ	rate [] "COVER SHEET FOR ASSIGNMENT (DOCUMENT) ACCOMPANYING NEW PATENT APPLICATION" or[] FORM PTO 1595 is also attached.	
14. [X] a. b. c. d.	Additional documents: [X] Copy of request (PCT/RO/101) [] International Publication No i. [] Specification, claims and drawing ii. [] Front page only [] Preliminary amendment (37 C.F.R. Section 1.121) [] Other	
15. [X] a. b.	The above checked items are being transmitted [X] before 30 months from any claimed priority date. [] after 30 months.	
16. []	Certain requirements under 35 U.S.C. 371 were previously submitted by the applicant on, namely:	

AUTHORIZATION TO CHARGE ADDITIONAL FEES

[X] The Commissioner is hereby authorized to charge the following additional fees that may be required by this paper and during the entire pendency of this application to Account No. 50-0850.

In re application of: Application No.:

ADDISON, P.S., et al. Not Yet Assigned

Group: Examiner: Not yet assigned Not yet assigned

(National Phase Entry of PCT/GB00/01675)

Filed:

Herewith

For:

METHOD OF ANALYSIS OF MEDICAL SIGNALS

[X] 37 C.F.R. Section 1.492(a)(1), (2), (3), and (4) (filing fees)

[X] 37 C.F.R. Section 1.492(b), (c) and (d) (presentation of extra claims)

[X] 37 C.F.R. Section 1.17 (application processing fees)

[] 37 C.F.R. Section 1.17(a)(1)-(5)(extension fees pursuant to Section 1.136(a). 37 C.F.R. Section 1.18 (issue fee at or before mailing of Notice of Allowance,

pursuant to 37 C.F.R. Section 1.311(b))

[] 37 C.F.R. Section 1.492(e) and (f) (surcharge fees for filing the declaration and/or filing an English translation of an International Application later than 30 months after the priority date).

Date: 01 November 2001

Customer No.: 26770

Respectfully submitted,

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1 "Method of Analysis of Medical Signals"

3 This invention relates to a method of analysis of

- 4 medical signals, and in particular to a method of
- 5 decomposition of cardiac signals using wavelet
- 6 transform analysis. Specifically the invention relates
- 7 to an improved method of resuscitation of patients in
- 8 cardiac arrest.
- 10 In the UK, coronary heart disease is the second
- 11 greatest contributor to deaths of people under 75. The
- 12 social and economic consequences of these death rates

1 are enormous. The current survivability rates of 2 patients after sudden cardiac failure are around 1:10. 3 4 Ventricular tachyarrhythmias, specifically ventricular 5 fibrillation (VF), are the primary arrhythmic events in 6 cases of sudden cardiac death. Administration of 7 prompt therapy to a patient presenting with such 8 symptoms can however lead to their successful 9 resuscitation. Until recently, the only indicators of 10 likelihood of survival of a patient to hospital 11 discharge were traditional variables such as emergency 12 service response time or bystander cardio-pulmonary 13 resuscitation (CPR). 14 15 In most cardiac complaints, analysis of a surface 16 electrocardiogram (EKG) of the presenting patient is a 17 rich source of information. However, until recently, a 18 surface EKG recorded during VF and any subsequent 19 medical intervention to defibrillate, was thought 20 merely to present unstructured electrical activity, and 21 not to provide useful information. 22 23 The first attempts to derive prognostic information from EKGs of the heart in VF focussed on the importance 24 25 of the amplitude of the waveform defined using peak-to-26 trough differences in the EKG voltage, measured as either the greatest deflection occurring in a 27 28 predefined time slot, or as the average peak-to-trough 29 voltage measured over a given time interval. 30 been shown that the VF amplitude is inversely related to time elapsed since collapse, is a crude predictor of 31 32 defibrillation outcome, and is a better indicator of

32

3

survival to hospital discharge than the traditional 1 variables described above. 2 3 However, recording the VF amplitude accurately is 4 significantly problematical. The EKG voltage amplitude 5 measured during VF is dependent on the direction of the 6 main fibrillation vector and is influenced by a variety 7 of factors including patient chest shape; electrode 8 9 size; electrode location; and skin/electrode interface resistance. This number of variables makes this 10 amplitude measure both unreliable and inaccurate. 11 is, although the amplitude of the waveform of an EKG 12 recorded during VF is now recognised to be a crude 13 predictor of the likely outcome of resuscitation of a 14 15 patient in VF, it is not a reproducible marker of sensitivity to defibrillation, and lacks clinical 16 17 usefulness. 18 In a further development, it is also known to use Fast-19 Fourier based transforms to generate a frequency 20 spectrum of an EKG in VF to analyse the signal. The 21 median frequency (MF) divides the area under the 22 spectrum into two equal parts. Since this plot is 23 derived from information in both the voltage and time 24 domains, external variables such as lead placement have 25 26 less effect on the results than the method of observing 27 the amplitude. However, CPR produces artefacts in the recorded EKG signal and, since pausing CPR merely to 28 obtain an EKG signal free of artefacts is likely to 29 compromise resuscitation, these artefacts are 30

necessarily included in this frequency measure, and

detract from its usefulness.

1	
2	Thus the results of such signal analysis show some
3	correlation with the likely outcome of resuscitation,
4	but again lack sufficient sensitivity and specificity
5	for clinical use. That is, this form of analysis has
6	the disadvantage that, since the Fourier spectrum
7	contains only globally averaged information, specific
8	features in the signal are lost.
9	
10	A method of accurate analysis of a surface EKG waveform
11	recorded during VF would therefore be useful in
12	understanding the pathophysiological processes in
13	sudden cardiac death, and thus to produce a model for
14	use:
15	
16	in predicting the efficacy of therapy in individual
17	cases; and
18	
19	in determining the selection of the preferred course of
20	primary, and alternative or adjunct therapies thus
21	providing a means for individually tailored therapy for
22	the specific patient needs
23	
24	to improve the success rate of resuscitation of
25	patients presenting in VF.
26	
27	Atrial fibrillation (AF) is a common cardiac arrhythmia
28	in older people. Atrial fibrillation can be stopped by
29	giving an electric shock to the patient under general
30	anaesthetic (cardioversion). However, many patient
31	return to an AF rhythm soon after treatment. The
32	technology detailed here may also provide a tool to

30 31

1	facilitate the clinical evaluation of AF exhibited in
2	the electrocardiogram (EKG) so reducing the risk
3	associated with general anaesthetic in patients where
4	the applied therapy is likely to prove ineffective.
5	
6	According to the present invention there is provided
7	a method of decomposition of waveforms in a cardiac
8	signal using wavelet transform analysis.
9	
10	The method of the invention is non-invasive, accurate,
11	and capable of delivering real-time information.
12	
13	Preferably said method employs discretized wavelet
14	transform analysis to process the EKG.
15	
16	Preferably said method employs discretized continuous
17	wavelet transform analysis to process the EKG.
18	
19	Preferably said method comprises the steps of deriving
20	the wavelet energy surfaces of an EKG signal; and
21	plotting said wavelet energy surfaces against a
22	location parameter b , and a scale parameter. The s cale
23	parameter may be dilation a or band pass frequency f_{bpc}
24	
25	The method initially comprises the steps of connecting
26	electrodes to the presenting patient; and sampling the
27	analogue input signal to derive the cardiac signal.
28	

Typically said method comprises the step of visually

displaying the cardiac signal.

- 1 Said method may display the distribution of energies
- 2 within the cardiac signal. Said method may display
- 3 coherent structures within the cardiac signal.

- 5 Said display may be by means of a contour plot. Said
- 6 display may be by means of a surface plot. Preferably
- 7 said method provides means to visualise the signal in
- 8 real-time for clinical use.

9

- 10 Preferably said method is applicable in the analysis of
- 11 an EKG in ventricular fibrillation.

12

- 13 Said method may be applicable in the analysis of an EKG
- in ventricular fibrillation after the commencement of
- 15 cardio-pulmonary resuscitation (CPR).

16

- 17 The method may include the step of disassociating the
- 18 component features of the temporal trace of a recorded
- 19 EKG. Additionally or alternatively said method may
- 20 include the step of temporal filtering of an EKG signal
- 21 of a heart which is subject to CPR to disassociate the
- 22 CPR signal from the heart signal.

23

- 24 Typically said method provides measurable
- 25 characteristics for the estimation of the health of a
- 26 heart in VF. Said method may provide measurable
- 27 characteristics for the estimation of the health of a
- 28 heart in AF. Said me may provide Typically said method
- 29 provides measurable characteristics for the estimation
- 30 of the health of a heart.

- 1 The method may provide measurable characteristics for
- 2 the estimation of the time elapsed since the onset of a
- 3 cardiac incident.

- 5 Typically said method provides measurable
- 6 characteristics for the estimation of the health of a
- 7 heart after commencement in CPR.

8

- 9 Said method may provide a prediction for the outcome of
- 10 a given therapeutic intervention and so aid the
- 11 clinical decision making process.

12

- 13 Said method may provide a basis for individual, patient
- 14 specific, protocols for therapeutic intervention.

15

- 16 The method may provide a guide to the optimal timing of
- 17 defibrillation of a heart in VF.

18

- 19 Said method may include the step of constructing a
- 20 damage index for reference purposes. Construction of
- 21 said index might involve the development of a network
- 22 classifier from a library of recorded data. Said
- 23 network classifier may comprise a neural network. Said
- 24 network classifier may comprise a wavelet network
- 25 classifier.

- 27 Application of the method of the invention represents a
- 28 significant advance in coronary care by providing a
- 29 reliable predictor of the outcome of shocking a patient
- 30 in VF. In addition, the development of an algorithm
- 31 using the method of the invention gives the ability to
- 32 predict shock outcome and to facilitate individual

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1 patient therapy. The ability to provide patient 2 specific therapeutic intervention is a priority in the 3 advancement of currently applied medical protocols. 4 That is, as discussed above, in certain instances, 5 after prolonged cardiac arrest preceding defibrillation 6 7 pharmacological measures or CPR can increase the chance 8 of successful resuscitation. Thus, employing the method to predict the outcome of shocking avoids futile 9 10 defibrillation attempts which can even harm the heart, 11 and can indicate the need for intervention, and 12 influence the selection of the preferred type of intervention, to optimise the metabolic state of the 13 14 heart prior to counter-shock. 15 The predictor algorithm developed using the method is 16 17 being tested using a new generation of defibrillation devices that have the flexibility to allow easy 18 19 prototyping of the new defibrillation algorithms. 20 21 According to a further aspect of the present invention 22 there is provided a method of decomposition of 23 waveforms in a cardiac signal using matching pursuit 24 algorithms. 25 26 According to a further aspect of the present invention 27 there is provided an apparatus for decomposition of 28 waveforms in a cardiac signal, said apparatus 29 comprising wavelet transform analysis means. 30 31

31 Said apparatus may include means to display the 32 distribution of energies within a waveform.

- 1 Said apparatus may include a monitor adapted to display
- 2 decomposed waveforms. Said apparatus may be adapted
- 3 for inclusion in an EKG apparatus.

- 5 According to a further aspect of the present invention
- 6 there is provided defibrillation means adapted to
- 7 operate in response to a signal generated by comparison
- 8 of an EKG trace with decomposed waveform.

9

- 10 That is, the invention preferably provides a method of
- 11 wavelet analysis of cardiac signals which provides
- 12 structural information about the heart whether the
- 13 heart is healthy or not and has significant
- 14 advantages over fast Fourier transforms.

15

- 16 The invention may provide a display device in the form
- 17 of a scrologram that provides real-time visualisation
- 18 of a wavelet scalogram, showing the distribution of
- 19 energies and coherent structures within the signal for
- 20 use as guidance by a clinician.

21

- 22 The invention may further provide a data analysis tool,
- 23 which assists in shock timing (atrial pulsing). That
- 24 is, the derived data may indicate the optimum time to
- 25 administer shock to the heart. The invention may
- 26 provide a damage index, preferably in the form of an
- 27 artificial neural network.

- 29 Preferably the invention provides dissociation of the
- 30 component features of a temporal trace of a cardiac
- 31 signal, which may for example be CPR, AF, or cardio-
- 32 phonographic signals.

-1	Embodiments of the invention will now be described by
1	
2	way of example only and with reference to the
3	accompanying drawings in which:
4	
5	
6	Figure la is a Mexican hat wavelet;
7	
8	Figure 1b is the real part of a complex Morlet
9	wavelet;
10	
11	Figure 2a is a schematic plot showing the dilation
12	of a continuous wavelet;
13	
14	Figure 2b is a schematic plot showing the
15	translation of a continuous wavelet;
16	
17	Figures 3a to Figure 3e are the plots of the
18	'investigation' of a sinusoidal signal by Mexican
19	hat wavelets of various sizes, showing the effect
20	of translation of the wavelet along the signal
21	(change in b), and dilation of the wavelet (change
22	in a);
23	
24	Figure 4a is the plot of five cycles of a sine
25	wave of period P;
26	
27	Figure 4b is the contour plot of $T(a,b)$ against a
28	and b for the sine wave of Figure 4a;
29	
30	Figure 4c is the isometric surface plot of $T(a,b)$
31	against a and b for the sine wave of Figure 4a;
32	

1	Figure 5a is the plot of a combination of two sine
2	waves of period P1, and P2, where P1 = $5P2$;
3	
4	Figure 5b is the contour plot of $T(a,b)$ against a
5	and b for the sine wave of Figure 5a;
6	
7	Figure 5c is the isometric surface plot of $T(a,b)$
8	against a and b for the sine wave of Figure 5a;
9	
10	Figure 6a is an EKG trace of a pig heart in sinus
11	rhythm;
12	
13	Figure 6b is a 2D energy scalogram associated with
14	the EKG trace of Figure 6a;
15	
16	Figure 6c is a 3D energy scalogram associated with
17	the EKG trace of Figure 6a;
18	
19	Figures 6d, 6e, 6f and 6g are the energy surface
20	plots from four segments of an EKG signal
21	subsequent to the onset of VF, showing the three
22	dominant ridges A, B, and C appearing in the
23	transform surface, and showing in Figure 6g the
24	onset of CPR after five minutes, associated with a
25	gradual increase in passband frequency of the
26	ridges A,B, and C;
27	
28	Figure 7a is an energy scalogram for a pig heart
29	for the first seven minutes of ventricular
30	fibrillation, indicating the initiation of CPR
31	after five minutes;

1	Figure 7b is a schematic diagram of the salient
2	features of the scalogram of Figure 7a;
3	
4	Figure 7c is the smoothed plot of energy at the
5	8Hz level in the scalogram of Figure 7a against
6	time;
7	
8	Figure 8a is a typical segment of an EKG trace of
9	a pig heart in VF;
10	
11	Figures 8b, 8c, and 8d are the energy scalograms
12	associated with the trace of Figure 8a;
13	
14	Figure 9 is a screen shot of a real time viewer
15	which shows the collected EKG data with its
16	associated wavelet energy display in the form of
17	its energy scalogram, where windows scroll to the
18	right;
19	
20	Figure 10a is a 7 second trace of human ECG
21	showing a shock event;
22	
23	Figure 10b is a scalogram corresponding to the
24	trace of Figure 10a;
25	
26	Figure 11a shows the proportion of energy in
27	scalograms for 120 results (60 ROSC, and 60
28	asystole) at 1.9 Hz after shocking;
29	
30	Figure 11b shows the proportion of energy in
31	scalograms for 120 results (60 ROSC, and 60
32	asystole) at 9.3 Hz after shocking;

1	
2	Figure 12a is a schematic representation of
3	overlapping signal segments used in a neural
4	network test study;
5	
6	Figure 12b shows the weights attributed by the
7	Kohonen network to the 30 frequency levels used in
8	the scalogram;
9	
LO	Figure 13a is an aorta pressure trace;
11	
12	Figure 13b shows the EKG for the same time period
13	as the trace of Figure 13a; and
L4	
15	Figure 13c is the scalogram associated with the
16	trace of Figure 13a derived from the Morlet
17	wavelet;
18	
19	Figure 13d is a detail of the phase part of
20	scalogram Figure 13c;
21	
22	Figure 13e is the scalogram associated with the
23	trace of Figure 13a derived from the Mexican hat
24	wavelet; and
25	
26	Figure 13f demonstrates the correlation of aorta
27	pressure pulse position with lines of zero phase;
28	
29	Figures 14a is the plot of an EKG trace. Figure
30	14b is its associated phase at around 1.5Hz.
31	Figure 14c is its energy scalogram. The
32	correlation of zero phase at this lower frequency

31

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1	and high frequency (low dilation) peaks is thus
2	illustrated.
3	
4	Figure 15a shows a 2 second segment of EKG taken
5	from a patient with atrial fibrillation (AF).
6	Figure 15b shows the wavelet scalogram plot
7	associated with this EKG. Figure 15c shows the
8	corresponding modulus maxima of the scalogram of
9	Figure 15b.
10	
11	Figure 15d contains a 7 second segment of EKG
12	exhibiting AF. Figure 15e is a trace of EKG
13	temporal components with small amplitude. Figure
14	15f shows the larger magnitude components i.e. the
15	QRS and T waves.
16	
17	Figure 15g is a plot of a two second 'blow up' of
18	part of the signal of Figure 15d; Figure 15h is a
19	plot of a two second 'blow up' of part of the
20	signal of Figure 15e; and Figure 15i is a plot of
21	a two second 'blow up' of part of the signal of
22	Figure 15f.
23	
24	Referring to the Figures, the present method employs
25	the use of a wavelet transform to analyse a cardiac
26	signal.
27	
28	The method involves the decomposition of the signal.
29	This decomposition is accomplished by utilising wavelet
30	transforms to decompose the signal in wavelet space.

1 A key distinction between the Fourier analysis of an EKG signal and its analysis by means of a wavelet 2 function is that, whilst the Fourier transform employs 3 a sinusoid function, a wavelet function is localised in 4 5 time. 6 7 The methodology for such decomposition may include discretized continuous wavelet transforms, orthonormal 8 wavelet transforms of decimated construction, non-9 decimated wavelet transforms, wavelet packet transforms 10 11 and matching pursuit algorithms. 12 Signal processing employing wavelet transform analysis 13 allows simultaneous elucidation of both spectral and 14 15 temporal information carried within a signal. Such processing can employ either continuous or discrete 16 The choice of wavelet transform used for a 17 particular signal processing application depends on 18 factors such as speed of computation necessary, the 19 shape of signal specific features, the frequency 20 resolution required, and the statistical analysis to be 21 22 performed. 23 The preferred method employs the discretized continuous 24 transform as it provides high resolution in wavelet 25 space at lower frequencies. 26 27 This method thus employs the use of a discretized 28 continuous wavelet transform to analyse a cardiac 29 30 signal.

- 1 In particular, this method employs a wavelet transform
- 2 as an interrogation tool for EKG signals of ventricular
- 3 fibrillation.

4

- 5 A variety of wavelet functions are available, and the
- 6 most appropriate is selected to analyse the signal to
- 7 be investigated.

8

- 9 The wavelet transform of a continuous time signal,
- 10 x(t), is defined as:

11

12 $T(a,b) = \frac{1}{w(a)} \int_{-\infty}^{\infty} x(t) \overline{g} \left(\frac{t-b}{a} \right) dt$ equation 1

13

- 14 where g(t-b)/a) is the analysing wavelet function and
- 15 '' denotes complex conjugate. w(a) is a scaling
- 16 function usually of the form $w(a) = a^n$ where n is usually
- 17 1 or 0.5, and x(t), in this application, is the single
- 18 channel surface EKG time signal. The transform
- 19 coefficients T(a,b) are found for both specific
- 20 locations on the signal, b, and for specific wavelet
- 21 dilations, a. T(a,b) is plotted against a and b in
- 22 either a surface or contour plot.

- 24 While other wavelet types may be employed the wavelets
- 25 mainly used in this method are: the Mexican hat wavelet
- 26 and the Morlet wavelet, examples of which are shown in
- 27 Figure 1.

- 1 The wavelet can translate along the signal (change in
- 2 b) and dilate (change in a). This is shown
- 3 schematically in Figure 2 using a Mexican hat wavelet.
- 4 Figure 3 illustrates the way in which a sinusoidal
- 5 signal can be 'investigated' at various locations by
- 6 Mexican hat wavelets of various sizes. The numerical
- 7 value of the convolution (equation 1) depends upon both
- 8 the location and dilation of the wavelet with respect
- 9 to the signal.
- 10 Figure 3a shows a wavelet of similar 'size' to the
- 11 sinusoidal waves superimposed on the signal at a b
- 12 location which produces a reasonable matching of the
- 13 wavelet and signal locally. From the Figure it is
- 14 apparent that there is a high correlation between the
- 15 signal and wavelet at this a scale and b location.
- 16 Here, the cross correlation of the signal with the
- 17 wavelet produces a large positive number T(a,b).
- 18 Figures 3b and 3c show details of the wavelet transform
- 19 of a signal using a wavelet of approximately the same
- 20 shape and size as the signal in the vicinity of b.
- 21 Figure 3b shows a wavelet of similar scale to the
- 22 sinusoidal waveform located at maximum negative
- 23 correlation. This produces a large negative T(a,b)
- 24 value. Figure 3c shows a wavelet of similar scale to
- 25 the sinusoidal waveform located at a position on the
- 26 time axis where near zero values of T(a,b) are
- 27 realised. Figure 3d shows the effect on the transform
- 28 of using the smaller a scale. It can be seen from the
- 29 plot that the positive and negative parts of the
- 30 wavelet are all in the vicinity of approximately the

- 1 same part of the signal, producing a value of T(a,b)
- 2 near zero. Figure 3e shows that the same thing happens
- 3 when using a much larger wavelet, since the wavelet
- 4 transform now covers various positive and negative
- 5 repeating parts of the signal, again producing a near
- 6 zero value of T(a,b).

- 8 Wavelet transforms are not usually computed at
- 9 arbitrary dilations for isolated locations in the
- 10 signal, but rather over a range of a and b. A plot of
- 11 T(a,b) versus a and b for sinusoidal data using the
- 12 Mexican hat wavelet is shown in Figure 4. Two methods
- are then employed to plot T(a,b), namely a contour plot
- or scalogram as shown in Figure 4b, and a surface plot
- 15 as shown in Figure 4c. At small and large values of a,
- 16 the near zero values of T(a,b) are evident from the
- 17 plots, but at values of a of the order of one quarter
- 18 of the wavelength of the sinusoid large undulations in
- 19 T(a,b) correlate with the sinusoidal forms of the
- 20 signal.

- 22 Figure 5a shows two superpositioned sinusoidal
- 23 waveforms, the first with period P1, the second with
- 24 period P2. P1 = 5P2. Figures 5b and 5c, the transform
- 25 plots of the superimposed waveforms clearly show the
- 26 two periodic waveforms in the signal at scales of one
- 27 quarter of each period. Thus, Figure 5 clearly
- demonstrates the ability of the continuous wavelet
- 29 transform to decompose the signal into its separate

- 1 frequency components. That is, this transform
- 2 'unfolds' the signal to show its constituent waveforms.
- 3 The contribution to the signal energy at a specific a
- 4 scale and b location is proportional to the two-
- 5 dimensional wavelet energy density function which is,
- in turn, proportional to the modulus of T(a,b).

7

- 8 The method of the present invention thus involves the
- 9 display of the transform as a contour plot. That is,
- 10 the method is used to present information derived from
- 11 an EKG trace of the heart in VF as a scalogram. The
- 12 preferred form of presenting the information is as an
- 13 energy scalogram, which presents the results as a plot
- 14 showing the log of the wavelet energy coefficients,
- 15 against the log of the bandpass centre frequency, f_{bpc} ,
- 16 of the wavelets for each time increment. The bandpass
- 17 centre frequency is proportional to the reciprocal of
- 18 the dilation value, a. This plot highlights small
- 19 changes in amplitude over the scales of interest. The
- 20 transform copes with repeating features in time with
- 21 shifting phase, making it appropriate for real time
- 22 applications such as this.

23

- 24 That is, by performing continuous wavelet transform
- 25 analysis on the ECG in VF, and then by producing an
- 26 energy scalogram of the results, it is possible to
- 27 unfold the signal in such a way that a previously
- 28 hidden structure is apparent, in contrast to the
- 29 apparently disorganised VF signal.

- The method then includes quantifying the wavelet 1 2 decomposition. This wavelet decomposition provides both qualitative visual and measurable features of the 3 EKG in wavelet space. 4 5 In practice, surface EKG tracings, recorded as soon as 6 7 possible after the onset of VF, are analysed. 8 As a demonstration of the efficacy of the method, in an 9 example of an experimental procedure utilising this 10 method of analysis employing wavelet techniques, VF was 11 induced in anaesthetised pigs via a pacemaker probe, 12 using a 90V impulse at 60 Hz. All of the pigs remained 13 in VF, untreated for a period of either 3 or 5 minutes. 14 15 After this time, CPR commenced. The surface EKG (standard lead II) was recorded using needle 16 electrodes. The EKG was sampled at 300 Hz using a 12-17 bit A to D converter. The method of the present 18 invention was then performed using 32 EKG tracings 19 recorded immediately after the onset of VF. 20 21 Figure 6a represents 4 beats of a pig heart in sinus 22 rhythm. Figures 6b and 6c shows the wavelet transform 23 24 of the signal displayed in two and three dimensions respectively. 25 26
- The QRS complex of the waveform is evident from the conical structures in Figure 6b converging to the high frequency components of the RS spike. The P and T
- 30 waves are also labelled in the plot. The 3D landscape
- 31 plot of Figure 6c shows the morphology of the signal in

- 1 wavelet space. In Figures 6b and 6c the continuous
- 2 horizontal band (X) is associated with a frequency of
- 3 1.7 Hz, the beat frequency of the sinus rhythm. The
- 4 second band (Y) occurs at a frequency of approximately
- 5 5.1 Hz, corresponding to the separation of the P-QRS-T
- 6 components in time. At higher frequencies the P, QRS
- 7 and T components are individually resolved according to
- 8 their frequency makeup and temporal location.

9

- 10 Figures 6d to 6g show the energy surfaces for four
- 11 segments of EKG signal subsequent to the onset of VF,
- 12 namely: (6d) 0-60 s; (6e) 60-100 s; (6f) 210-240 s;
- 13 and (6g) 260-360 s.

14

- 15 The morphology of the VF signal in wavelet space can be
- 16 seen from the Figures to contain underlying features
- 17 within a more complex surface topography. The most
- 18 significant features are the dominant ridges that
- 19 appear in the transform surface through time.

20

- 21 Figure 6f shows these ridges quite clearly. A high-
- 22 energy ridge can be observed at around 10 Hz and two
- 23 lower energy bands can be observed at lower
- 24 frequencies. These three ridges are labelled A, B and
- 25 C, respectively, in the plot. Other ridges are also
- 26 present within the scalogram.

- 28 The energy surface in Figure 6g contains the onset of
- 29 CPR after 5 min of untreated VF. The institution of
- 30 CPR is associated with a gradual increase in the
- 31 passband frequencies of ridges A, B and C. This change
- 32 in the composition of the VF signal reflects electrical

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1 changes in the fibrillating myocardium associated with 2 the onset of CPR. This is because CPR produces antegrade myocardial blood flow and thus improves the 3 metabolic state of the tissues, temporarily reversing 4 the otherwise progressive decline in high band pass 5 6 frequency components of the EKG wavelet decomposition. 7 Figure 8a is a typical segment of an EKG trace of a pig 8 9 heart in VF; Figures 8b, 8c, and 8d are the energy scalograms associated with the trace of Figure 8a. As 10 11 clearly illustrated by these diagrams the principle 12 dilation (band pass centre frequency) component of the scalogram is approximately 10Hz. However, using said 13 14 method it is also apparent that this component is not 15 constant. It 'pulses' with a degree of regularity. This 16 structure is previously unreported. 17 18 Figure 9 shows similar 'pulsing' in another porcine EKG 19 signal. However, the structure is so pronounced that 20 high energy, high frequency, intermittent components can be observed. These components have an occurrence 21 22 frequency of the order of the original sinus rhythm: 23 approximately 1.7Hz. 24 25 Figure 10a is a human EKG signal segment containing a shock event. Figure 10b is the corresponding wavelet 26 27 scalogram. It is apparent from the scalogram of Figure 10b that both high frequency spiking and an 28

intermittent high-energy region are present in the

vicinity of 10 Hz and also above 10Hz.

30 31

- 1 The high frequency spiking is unique to the method of
- 2 the present invention and is not visible using
- 3 conventional Fourier techniques. The rich structure
- 4 made visible within the EKG by the wavelet transform
- 5 method is evident in the scalogram.
- 6 It is clearly seen from the Figures that applying the
- 7 wavelet transform to an EKG signal of VF demonstrates
- 8 that this signal is a rich source of valuable
- 9 information. That is, it produces a display showing
- 10 real time visualisation of the distribution of energies
- 11 and coherent structures within the signal for use by a
- 12 clinician in the selection of treatment strategies.
- 13 Using this method of analysis it is feasible to obtain
- 14 real-time visual display of the EKG frequency
- 15 characteristics in the wavelet domain during
- 16 resuscitation. The scalogram produced provides
- 17 information about the myocardium that is not available
- 18 from a standard single channel surface EKG.
- 19
- 20 The wavelet scalogram decomposition can be displayed as
- 21 a real time scrolling window, as shown in Figure 9.
- 22 This window is useful as an aid for clinical decision
- 23 making. It can be used as a stand-alone tool, or as
- 24 basis for on-line statistical analysis of the current
- 25 state of a heart.
- 26
- 27 To produce the window, a MATLAB TM R11 application is
- 28 used. Each EKG sample taken results in the updating of
- 29 a FIFO (First In First Out) buffer, and the EKG plot of
- 30 Figure 9a. The scalogram of Figure 9b is then shifted

to the right and clipped before the 'missing' new right hand data is calculated, using conventional matrix algebra, and filled.

5 This results in the two scrolling windows of Figure 9.

- 6 The exponential ramp in the bottom right corner shows
- 7 the compact support of the wavelet utilised at the
- 8 given scale.

9

4

- 10 Higher resolution scalograms are achieved through
- implementation on higher specification machines,
- 12 purpose built hardware, or application specific
- 13 software with coding using a lower level programming
- 14 language, such as C++.

15

- 16 CPR produces artefacts in the EKG signal. Additionally,
- 17 this method delivers information the value of which is
- 18 not degraded once the CPR artefacts are filtered from
- 19 the EKG signal.

20

- 21 From examination of the scalograms shown in Figures 6g,
- 22 7a and 7b it can be seen that the VF signature and the
- 23 signature of the CPR artefacts occupy distinct areas of
- 24 the scalogram, which permits their separation.

- 26 Known techniques such as the Modulus maxima method are
- 27 now available to reduce the non-zero data points in the
- 28 wavelet scalogram. This method reduces the topography
- 29 of the scalogram surface to a series of ridges, thereby

1 considerably reducing the amount of data required to

2 represent the signal in the wavelet space.

3

4 The modulus maxima obtained from a bandlimited signal

- 5 with a wavelet of finite compact support in the
- 6 frequency domain defines a complete and stable signal
- 7 representation.

8

- 9 In this method, temporal filtering of the original EKG
- 10 signal to disassociate the CPR signature from the heart
- 11 signal can either be done directly, using the wavelet
- 12 energy scalograms, or indirectly through modulus maxima
- 13 techniques. This allows the heart to be monitored
- 14 without necessitating cessation of CPR to allow rhythm
- 15 recognition.

16

- 17 Further to the above, the method may also be applied to
- 18 patients suffering form atrial fibrillation (AF) as a
- 19 means of disassociating the prevalent QRS and T waves
- 20 from the remainder of the signal.

21

- 22 Wavelet decomposition of the ECG signal is performed
- 23 using an appropriate wavelet function. The modulus
- 24 maxima technique is used to encapsulate the scalogram
- 25 information in a series of ridges. Filtering of the
- 26 signal is then undertaken using the modulus maxima
- 27 information and through reconstruction the clinically
- 28 useful information is isolated from the signal .

- 30 Specifically, Figure 15a shows the wavelet transform
- 31 decomposition of a 2 second segment of ECG taken from a
- 32 patient with atrial fibrillation. Below the ECG trace

lars.

1 is a wavelet scalogram plot. The corresponding modulus

2 maxima of the scalogram is plotted below the scalogram.

3

For example, Figure 15d contains a 7 second segment of 4 ECG exhibiting AF. The signal has been partitioned 5 using a modulus maxima ridge following algorithm. The 6 7 modulus maxima ridges have been separated into large and small scale features by thresholding the signal at 8 9 a predetermined wavelet scale. A blow up of part of the signal is given in the lower three plots in the figure: 10 11 Figures 15q, 15h and 15i. The middle of these plots contains the partitioned signal with the QRS complex 12 and T wave filtered out revealing regular, coherent 13 14 features that appear at a frequency of approximately 15 400 beats per minute, typical of AF. The lower plot contains the partition with the filtered out QRS and T 16 waves. Although, a relatively simple modulus maxima 17 technique was used in this pilot study whereby the 18 modulus maxima lines were simple partitioned into two 19 subsets, the ability of the technique to separate the 20 21 signal into QRS and T waves and underlying atrial activity is evident from the results. 22 It is known that 23 the decay in amplitude ο£ a modulus maxima corresponding to a signal feature can be a function of 24 the scale of the wavelet. It is possible to use this 25 property to separate the ridge coefficients into a 26 In this further 27 noisy and coherent part. way, differentiation of the modulus maxima information can 28 be implemented within a more sophisticated algorithm. 29 facilitate the further separation 30 will 31 background noise, QRS and T waves, and atrial activity.

1 This method thus facilitates useful interpretation of

2 previously unintelligible EKG signals.

3 In patients presenting with uncoordinated rapid

4 electric activity of the ventricle of heart, known as

5 ventricular fibrillation (VF), there is no effective

6 pulse and myocardial blood flow ceases. Even the

7 institution of optimal cardio-pulmonary resuscitation

8 (CPR) of the patient does not achieve more than 30% of

9 the normal cardiac output. Ischaemia during cardiac

10 arrest leads to a rapid depletion of myocardial high-

11 energy phosphates, deterioration of transmembrane

12 potentials, and disruption of intracellular calcium

13 balance. Paradoxically, the myocardium in VF has

14 supranormal metabolic demands. For this reason

15 resuscitation attempts become less likely to succeed

16 with the passage of time, and electrical defibrillating

17 shocks increasingly result in asystole or EMD.

18

19 After prolonged cardiac arrest, the use of

20 pharmacological measures or CPR before attempting

21 defibrillation may increase the chances of successful

22 resuscitation. This invention provides a robust and

23 reliable method of analysis of the state of the

24 myocardium in VF that prevents attempts to defibrillate

25 at times that are unlikely to be successful, or even

26 harmful to the heart. This method also provides an

27 indication of the best way in which to optimise the

28 metabolic state of the heart prior to counter-shock.

28

The method includes steps to establish a standard 1 against which to evaluate collected data in a 2 particular incidence. 3 4 The method further employs use of measurable signal 5 characteristics derived from the position and amplitude 6 of features in the scalogram to estimate both the 7 condition of the myocardium, and downtime of the 8 subject while in VF. 9 10 The method thus provides for optimal treatment of the 11 heart in VF, so fulfilling specific patient needs, by 12 therapeutic intervention, if appropriate. 13 14 An energy scalogram such as that shown in Figure 7 15 displays three distinct bands, labelled A, B, C. It is 16 possible to derive quantifiable measures using 17 correlations between the location and energy content of 18 the bands. 19 20 Band A of Figure 7b represents the dominant energy band 21 seen in the scalogram of Figure 7a, and corresponds to 22 23 the tachycardic beating of VF. However the scalogram is much more informative in that it also shows, as 24 bands B and C, the behaviour of other frequency 25 components of the signal which were previously 26 27 unreported. 28 Figure 7a shows a 2D energy scalogram. It includes the 29 first 5 minute period of VF, followed by a 2.5 minute 30 period of CPR. The onset of CPR is clearly identified 31 by the distinct horizontal dark band in the lower right

- 1 quadrant of the Figure. Over the first 5 minute 2 period, three bands, labelled A, B, C, can be clearly
- 3 seen in the scalograms. These bands correspond to the
- 4 ridges of Figures 6d to q. The increase in the
- 5 frequency components of these three bands after the
- 6 onset of CPR is evident in the plot. Bands B and C
- 7 follow trajectories similar to each other in the
- 8 scalogram, reducing in frequency over time. Band A,
- 9 however, moves independently of the other two.
- 10 Initially Band A increases, then it decreases to a
- 11 local minimum value at approximately 70s. Between 70
- 12 and 160s it increases relative to Bands B and C.
- 13 Finally, it decreases until the start of CPR after
- 14 300s. The same pattern was present in all 32 pig EKG
- 15 traces of the experiment.

17 Obvious increases in the passband frequency of all

- 18 three bands are observed in the scalogram after the
- 19 onset of CPR. For some of the signals studied this
- 20 increase in band C is masked by the dominant CPR band,
- 21 and thus cannot be seen in the scalogram.

22

- 23 Figure 7b provides a schematic diagram of the salient
- 24 features contained within the scalogram plots, where t0
- 25 is immediately after the onset of VF; t2 is the start
- of CPR; and t3 is the end of the analysis. Figure 7c
- 27 shows the relative proportion of energy contained in
- 28 the scalogram in the 5 to 12 Hz region through time.
- 29 There is an obvious decay in the relative energy
- 30 associated with this region which is associated with
- 31 the breakdown of co-ordinated activity in the heart.

- 1 The steps of the method of the present invention
- 2 described above establish that during the course of VF
- 3 there is a reduction in the proportion of energy within
- 4 the dominant frequency band indicated in Figure 7c.
- 5 This dominant frequency band, Band A in Figure 7a, is
- 6 demonstrated to be approximately 10 Hz for pig VF.

7

- 8 The energy within this band changes rapidly. This is
- 9 illustrated by the 'pulses' in Figures 8,9,10.

10

- 11 The Figures 6,7,8,9,10 show that applying the wavelet
- 12 transform to an EKG signal of VF demonstrates that this
- 13 signal is a rich source of valuable information.

14

- 15 The underlying hypothesis of the method of the present
- 16 invention is that the scalogram associated with an EKG
- 17 correlates to the state of the myocardium as it decays
- 18 subsequent to the onset of VF.

19

- 20 The method uses the information contained in the energy
- 21 scalogram associated with an EKG to predict the likely
- 22 success of clinical intervention, namely shocking.

23

- 24 It is therefore possible to develop a wavelet transform
- 25 based tool for the prediction of shock outcome during
- 26 ventricular fibrillation by:

27

- 28 1. collecting and collating data from sets of
- 29 archived EKGs recorded from humans in VF where
- 30 attempts to resuscitate by shocking were made; and

31

32 2. developing a classifier for reference purposes.

1 Figure 11 is a classification of the shock outcome in 2 either asystole or a rhythmic response using a 3 relatively simple statistical analysis. The experiment 4 yielding the results to compile these Figures involved 5 use of the lead II outputs of standard three lead EKGs 6 of 120 patients in VF. Each trace is of three second 7 duration sampled at 100 Hz. Of these patients, 60 8 returned to sinus rhythm while the other 60 9 deteriorated to asystole, post shock. 10 11 Each trace was decomposed into an associated wavelet 12 transform from which its energy scalogram was 13 generated. The volume under this surface was then 14 normalised to render the results independent of signal 15 amplitude, but instead the result of the relative 16 wavelet constituents of the signals. The log of the 17 mean values at each dilation (band centre frequency) 18 for each was then recorded. Figures 11a and 11b show 19 the distribution of energies in a lower frequency band 20 (1.9 Hz) and at the 9.3 Hz band. Clearly, through 21 visual inspection, it is apparent that the proportion 22 23 of energies around the 10 Hz band is higher for successful defibrillation attempts. 24 25 The method then extends to apply neural techniques to 26 analysis of wavelet pre-processed EKG signals. 27 28 A pilot study conducted to determine the feasibility of 29 using artificial neural techniques to provide a tool to 30 predict the outcome of defibrillation during VF used 31

eight human EKG trace segments containing shock events.

In these cases, the result of shocking was unequivocal
four patients returned to VF, and four experienced

3 return of spontaneous circulation (ROSC).

4

5 The traces were transformed using the Morlet wavelet,

6 and energy scalograms containing thirty frequency

7 levels were produced. This was then split into eight

8 overlapping sections as shown in Figure 12a, each of

9 200 points (2/3 seconds duration). These 200 location

10 points were subsampled down to 50 to give eight

11 scalograms for each trace of 50 x 30 elements. The

12 volume under the energy scalograms were normalised and

13 the patterns fed into a 'winner take all' Kohonen

14 network with two output units and built in conscience

15 (to avoid local minima). That is, the network was

16 asked to group the 64 input patterns into two classes.

17 All but ten outputs were collectively classified

18 correctly giving a mean pattern error of 0.156 (against

19 0.5 average pattern error expected from random inputs).

20

21 Since this is a vector quantisation method (VQM) it was

22 possible to identify how the network differentiates the

23 patterns through inspection of its connective weights.

24 The weights from each location position across all

25 scales in the network are approximately the same, which

26 means that there are no markers with which to

27 synchronise the different pre-processed traces. This

28 confirms that this neural network is too simple for

29 this purpose. That is the network is not equipped to

30 'consider' the relative phase of each input pattern.

1	Figure 12b sh	ows the weights for the	'success' (ROSC)
2	and 'failure'	(VF) to the output unit	s from the first
3	two time slic	es across all scales. T	he weights
4	indicate the	classes are differentiat	ed by the
5	proportion of	energy in the lower sca	les, which can be
6	seen when com	pared with Figure 11.	
7			
8	Although the	above described method i	ndicates the
9	slight drop i	n the dominant frequency	expected, the
10	drop is very	marginal which leads to	the conclusion of
11	the lack of o	ompetence of previously	proposed methods
12	as a defibril	lation success predictor	: .
13			
14	In summary, a	library of human ECG da	ata containing data
15	sets of humar	VF with attempts to res	suscitate by
16	shocking is u	used as a database. This	s database is
17	extended to	nclude data sets contair	ning various
18	methods of sl	nocking including, for ex	kample, biphasic
19	shocking. The	ne biphasic shock wavefor	rm has resulted in
20	an increased	proportion of successful	l defibrillation
21	attempts and	is set to become the sta	andard treatment
22	for cases of	VF.	:
23			
24	In one examp	le, the recognised outcor	mes are defined by
25	trace compone	ents of the post-shock w	indow lasting until
26	next shock (if present). If the rat	io of the given
27	rhythm excee	ds 10% of the total windo	ow length the
28	rhythms are	prioritised according to	the sequence:
29			
30	Class	Rhythm	Ratio
31			
32	1	Pulse (SVR)	+10%

1	2	No pulse (E	MD)	+10%
2	3	Isoelectric	(Asystole)	+10%
3	4	VF		+10%
4				
5				
6	Class 5 is the	class of VF	preceding shocks	where VF
7	re-establishes	s itself with	in 5 seconds foll	owing the
8	shock (i.e. no	change). T	he VF in all the	other
9	classes were r	on-VF in thi	s period.	
10				
11	Wavelet analys	sis of this i	nformation in acc	cordance with
12	the method of	the inventio	n is then perform	med to:
13				
14	construct a wa	avelet visual	isation of the si	ignal -
15	usually by plo	otting wavele	t energy surfaces	s against the
16	location param	meter b and t	he inverse of the	e dilation
17	parameter a;			
18				
19	provide measu:	rable charact	eristics of the s	signal for
.20	estimation of	downtime of	the patient;	
21				
22	provide measu	rable charact	eristics of the	signal for
23	determining t	he health of	the heart post C	PR; and
24				
25	to construct	energy scalog	gram devised for	the method -
26	which uses th	e energy dens	sity function and	the
27	reciprocal of	the wavelet	a scale for use	as a
28	predictor too	1.		
29				
30	As described	above it is p	possible to use a	rtificial
31	neural networ	k based techr	niques to develop	such an
32	indication of	the state of	f myocardium. In	the

- 1 alternative, it is possible to classify the wavelet
- 2 scalogram through multilayered feedforward network
- 3 types.

- 5 The method may include the development of a modulus
- 6 maxima algorithm tool for the preprocessing of ECG
- 7 prior to its input into a neural network classifier.

8

- 9 Using this technique improves network performance
- 10 whether this data is further encoded, or presented as a
- 11 whole, larger, sparse matrix as a pattern in the input
- 12 space.

13

- 14 This method therefore utilises the generalisation
- 15 properties of a feed forward multi-layer network to
- 16 predict the likelihood of defibrillation success from
- 17 the wavelet transform of the EKG traces. This multi-
- 18 layer network with its relatively simple dynamics, when
- 19 combined with wavelet pre-processing, has proved itself
- 20 a useful tool as a universal approximator.

21

- 22 The classes of multi-layer network types of use in this
- 23 method are:

- Multi-layered feed forward (MLFF) neural networks
- 26 with back propagation training and monotonic
- 27 activation functions; and
- 28 Radial Basis Neural Networks (RBNN) as have
- 29 previously been successfully applied to the denoising
- 30 of medical Doppler ultrasound signals with wavelet
- 31 preprocessing.

1	
2	As described above, the method involves the
3	decomposition of EKG signals into a complete basis set
4	defined by the wavelet shape and other parameters by
5	salient basis functions of a different basis set,
6	converged upon through regression techniques (sigmoid
7	in the case of multilayer neural networks, Radial basis
8	etc).
9	
10	These regression techniques can also be used to
11	construct a wavelet basis function set directly.
12	
13	Methodologies for restricting the search space of the
14	wavelet basis functions considered are known. Whilst
15	this wavelet network has been shown to be effective for
16	chaotic time series prediction, its implementation
17	involves the use of wavelet frames of a decimated,
18	dyadic, construction. The method of the present
19	invention may employ continuous wavelet networks
20	spanning a redundant wavelet basis which, although
21	computationally more expensive, overcomes the time
22	invariance constraint and the limited size of input
23	space associated with use of wavelet frames.
24	
25	The method may use conventional gradient decent methods
26	to produce a single layer wavelet classifier.
27	
28	These wavelet networks may be further employed as part
29	of a multilayer system as a non-parameterised estimate
30	of the original trace for input to further hidden
31	layers.

32

37

The network type of choice for the automated prediction 1 system of the method is selected on the basis of its 2 sensitivity and selectivity in correctly classifying 3 successful defibrillation outcomes in test set data, 4 since this is most clinically useful. 5 6 Thus experimental comparison of the three techniques 7 demonstrates the efficacy of the wavelet transform 8 9 technique. 10 The nature of underlying atrial activity can also be 11 determined from wavelet decomposition of the EKG 12 The wavelet function gives information 13 signal. regarding the amplitude and, where appropriate, phase 14 of the transformed signal. It is known that pressure 15 readings taken from the aorta correlate to forms of 16 atrial activity within the heart. Areas of localised 17 high energy contained within the scalogram can be 18 demonstrated to correlate with these pressure readings. 19 This experimental result is extrapolated to mean that 20 areas of localised high energy contained within the 21 scalogram correlate with forms of atrial activity 22 23 within the heart. 24 Figure 13a shows the aorta pressure, Figure 13b the EKG 25 trace, for the same time period as Figure 13a, and 26 Figure 13c shows the scalogram for the EKG of Figure 27 It is apparent that there is an increase in 28 energy in the system during an atrial pulse, indicated 29 by the dark blotches occurring in the scalogram at an 30

 $f_{\rm bpc}$ of around 10 Hz. There is a frequency component

between 1 and 2 Hz. As shown in Figure 13d, which

- 1 highlights the phase of the scalogram between 1 and 2
- 2 Hz, it is apparent hat the lines of zero phase are in
- 3 alignment with the atrial pulse.

- 5 In a further scalogram, shown in Figure 13e, produced
- 6 by using the Mexican hat wavelet transform which is
- 7 real and has better temporal resolution, but worse
- 8 frequency resolution than the complex scalogram of
- 9 Figure 13c, it is demonstrated that positive high
- 10 amplitude components are shown at the same positions
- 11 for scales of between 1 and 2 Hz, thus reinforcing the
- 12 findings extrapolated from Figure 13c. That is as
- 13 shown in Figure 13f, the lines of zero phase correlate
- 14 with the pulse position.

15

- 16 The lines of zero phase within the 1.8Hz frequency band
- 17 also align with regular peaks in the scalograms, as
- 18 shown in Figures 14a, 14b & 14c. This links the
- 19 presence of the 1.8 Hz band with the observed peaks at
- 20 higher frequencies. This correlation between the 1.8
- 21 Hz band and the aorta pressure pulse suggests atrial
- 22 activity is present.

- 24 In a further application of the method, means for
- 25 identifying the optimum timing for application of the
- 26 defibrillation shock can be extrapolated from the
- 27 pulsing identified by the wavelet technique and shown
- 28 in Figures 8, 9, 10, and 14, by comparison with traces
- 29 of attempts at defibrillation which initially fail but
- 30 are subsequently successful.

1 Thus, any data sets, in the above, that correspond to 2 3 multiple shocking of the same patient, where defibrillation has been repeatedly attempted are 4 considered separately since these traces hold important 5 6 information. 7 The pilot study detailed above used Morlet wavelet 8 based energy scalogram decomposition of signal segments 9 immediately prior to shocking. A full parametric 10 11 wavelet study of the method determines the optimum 12 method. 13 14 The method includes the development of a classifier 15 using the wavelet transform analysis. 16 Various types of neural network classifier are 17 achievable using this method. 18 19 The linkage of shock timing to the phase information of 20 21 wavelet components allows for increased defibrillation success and reduced shock energies. The wavelet-22 23 derived information can also be employed to predict the 24 likelihood of shock success, preventing futile or 25 harmful defibrillation attempts, and providing a 26 predictor of an optimal resuscitation strategy or 27 strategies. 28 This method demonstrates the utility of the wavelet 29 30 transform as a new method of EKG signal analysis during 31 It provides a robust, real-time solution to the

problem of useful monitoring of the myocardium during
resuscitation.

3

- 4 When compared with conventional statistical methods,
- 5 such as fast Fourier transforms, it is seen that the
- 6 temporal resolution of the wavelet technique gives a
- 7 scalogram which better describes the non-stationary,
- 8 intermittent, nature of the EKG trace to be analysed,
- 9 and gives a method of greater predictive effectiveness
- 10 than is already known. The effectiveness criteria for
- 11 the networks of the method of the present invention are
- 12 based upon their sensitivity and selectivity in
- 13 correctly classifying successful defibrillation
- 14 outcomes from test data sets.

15

- 16 Although this description refers to wavelet transform
- 17 analysis, this term is to be construed to include
- 18 matching pursuit algorithms and similar analysis
- 19 techniques.

- 21 Modifications and improvements can be made to the above
- 22 without departing from the scope of the invention.

1	CLAI	MS
2		
3	1.	A method of decomposition of waveforms in a
4		cardiac signal using wavelet transform analysis.
5		
6	2.	A method as claimed in Claim 1 comprising the step
7		of employing discretized wavelet transform
8		analysis to process the said waveform.
9		
10	3.	A method as claimed in Claim 1 comprising the step
11		of employing discretized continuous wavelet
12		transform analysis to process the cardiac
13		waveform.
14		
15	4.	A method as claimed in any preceding claim
16		comprising the steps of deriving the wavelet
17		energy surfaces of an electrocardiogram (EKG)
18		signal; and plotting said wavelet energy surfaces
19		against a location parameter b , and a scale
20		parameter.
21		
22	5.	A method as claimed in Claim 4 wherein said scale
23		parameter is dilation a.
24		
25	6.	A method as claimed in Claim 4 wherein said scale
26		parameter is band pass frequency $f_{bpc}.$
27		
28	7.	A method as claimed in any preceding claim
29		comprising the initial steps of connecting
30		electrodes to a presenting patient; and sampling
31		the analogue input signals recorded to derive the

cardiac signal.

1	8.	A method as claimed in any preceding claim
2		including visually displaying the cardiac signal.
3		
4	9.	A method as claimed in any preceding claim
5		including visually displaying the distribution of
6		energies within the cardiac signal.
7		
8	10.	A method as claimed in any preceding claim
9		including visually displaying coherent structures
10		within the cardiac signal.
11		
12	11.	A method as claimed in any preceding claim
13		including visually displaying the signal in real-
14		time for clinical use.
15		
16	12.	A method as claimed in any preceding claim
17		comprising the step of constructing a contour plot
18		to display the decomposed waveform obtained.
19		
20	13.	A method as claimed in any preceding claim
21		comprising the step of constructing a surface plot
22		to display the decomposed waveform obtained.
23		
24	14.	- -
25		comprising the step of constructing a 2D or a 3D
26		energy scalogram to display the decomposed
27		waveform obtained.
28		
29	15.	• •
30		including the step of disassociating the component
31		features of the temporal trace of a recorded EKG.

1	16.	A method for the analysis of an EKG of a heart in
2		ventricular fibrillation including the method as
3		claimed in any preceding claim.
4		
5	17.	A method for the analysis of an EKG of a heart in
6		ventricular fibrillation after the commencement of
7		cardio-pulmonary resuscitation (CPR) including the
8		method as claimed in any of Claims 1 to 15.
9		
10	18.	A method as claimed in Claim 17 including the step
11		of temporal filtering of the EKG signal of a heart
12		that is subject to CPR to disassociate the CPR
13		signal from the heart signal.
14		
15	19.	A method as claimed in Claim 17 or Claim 18 using
16		wavelet energy scalograms.
17		
18	20.	A method as claimed in Claim 17 or Claim 18 using
19		ridge following techniques
20		-
21	21.	1
22		following techniques are modulus maxima
23		techniques.
24		
25	22.	
26		heart in VF including the method of any of Claims
27		1 to 15 to provide measurable characteristics.
28		
29	23.	
30		measurable characteristics are used to provide an
31		estimate of the time elapsed since the onset of a
32		cardiac incident.

1	24.	A method as claimed in Claim 22 wherein said
2		measurable characteristics are used to provide an
3		estimate of the health of a heart after
4		commencement of CPR.
5		•
6	25.	A method as claimed in any of Claims 22 to 24
7		wherein said measurable characteristics are used
8		to predict the outcome of a given therapeutic
9		intervention.
LO		
L1	26.	A method as claimed in any of Claims 22 to 25
L2		wherein said measurable characteristics are used
L3		to provide a guide for the optimal timing of
14		defibrillation of a heart in VF.
15		
16	27.	A method for the analysis of an EKG of a heart in
17		atrial fibrillation including the method as
18		claimed in any of Claims 1 to 14.
19		
20	28.	A method as claimed in Claim 27 including the step
21	1	of partitioning the signal to provide separate
22	•	traces of QRS and T waves, and/or atrial activity
23		and/or background noise.
24		
25	29.	A method as claimed in any preceding claim
26		including the step of constructing a damage index
27		for reference purposes.
28		
29	30.	A method as claimed in Claim 29 wherein
30		construction of said index includes the step of
31		developing network classifier from a library of
32		recorded data.

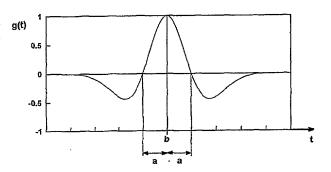
		45
1	31.	A method as claimed in Claim 30 wherein said
2		network classifier developed is a neural network.
3		
4	32.	A method as claimed in any of Claims 29 to 31
5		wherein said network classifier developed is a
6		wavelet network classifier.
7		
8	33.	A method of decomposition of cardiac waveforms
9		using matching pursuit algorithms.
.0		
L1	34.	Apparatus for decomposition of waveforms in a
L2		cardiac signal, said apparatus comprising wavelet
L3		transform analysis means.
14		
15	35.	Apparatus as claimed in Claim 34 including means
16		to display the distribution of energies within a
17		waveform.
18		
19	36.	Apparatus as claimed in Claim 34 or Claim 35
20		including a monitor adapted to display decomposed
21		waveforms.
22		
23	37.	- -
24		adapted for inclusion in an EKG apparatus.
25		
26	38.	
27		response to a signal generated by comparison of an
28		EKG trace with decomposed waveform obtained by the

method of any of Claims 1 to 33.

1	39.	A method as described in any of Claims 1 to 33
2		with reference to or as shown in the accompanying
3		drawings.
4		
5	40.	Apparatus as described in any of Claims 34 to 38
6		with reference to or as shown in the accompanying
7		drawings.
8		
9		
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11		
12		•
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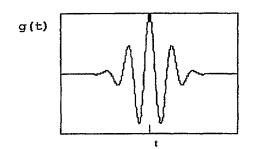
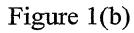
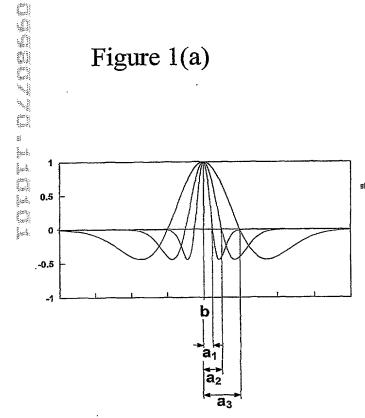


Figure 1(a)





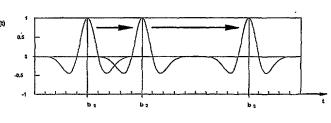
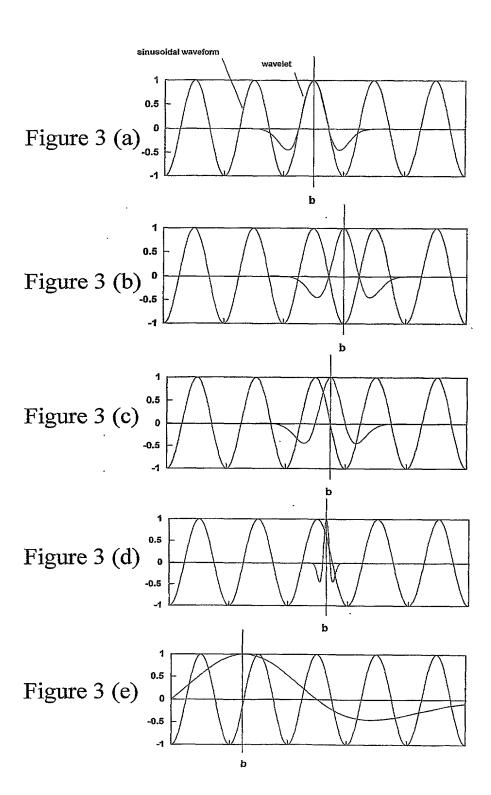


Figure 2(a)

Figure 2(b)



3/14

Figure 4 (a)

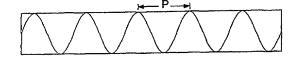


Figure 5 (a)

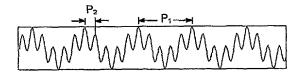


Figure 4 (b)

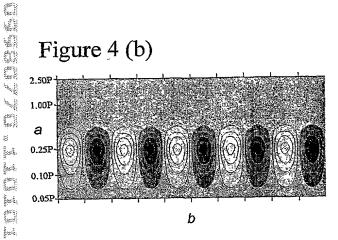


Figure 5 (b)

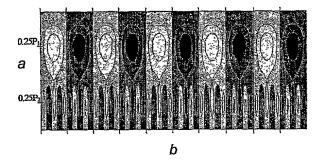


Figure 4 (c)

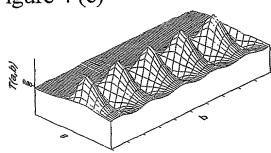


Figure 5 (c)

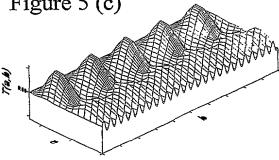


Figure 6 (a)

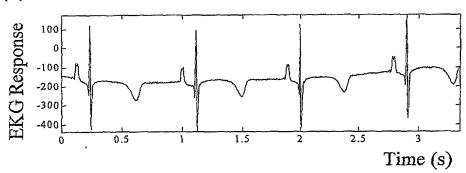


Figure 6 (b)

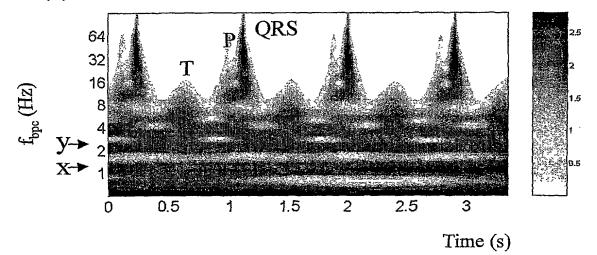
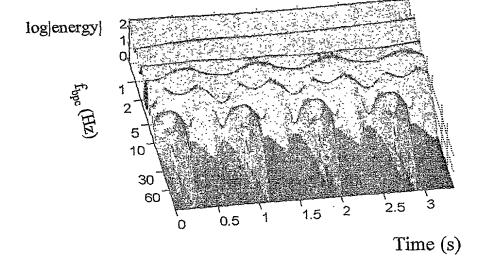
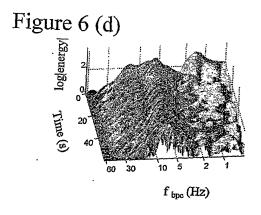
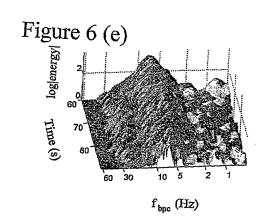
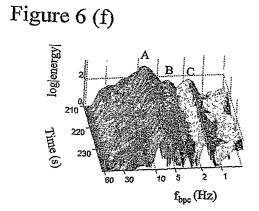


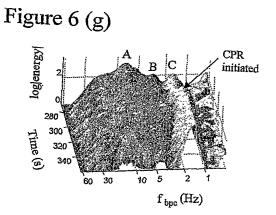
Figure 6 (c)











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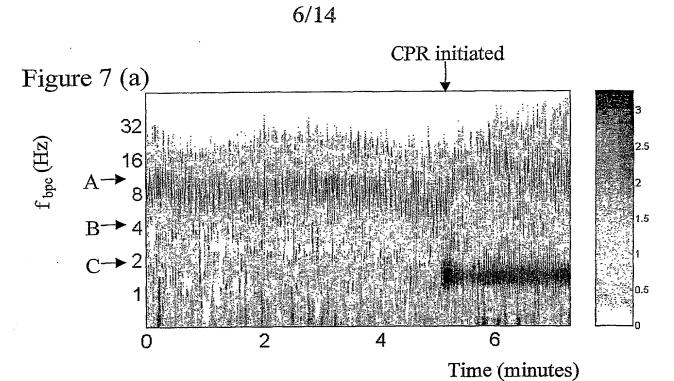


Figure 7 (b)

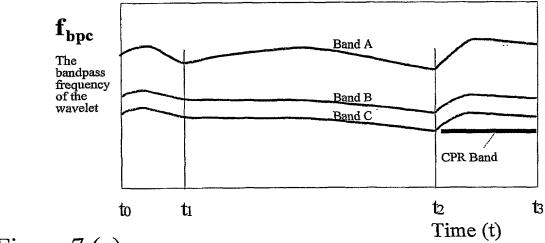
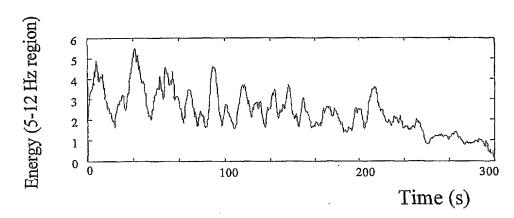


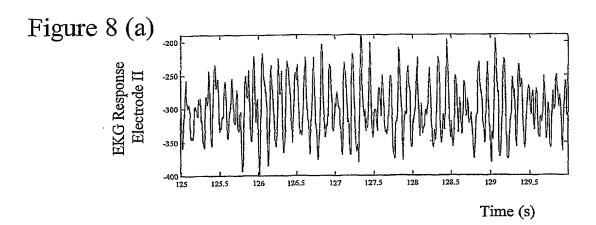
Figure 7 (c)



129.5

Time (s)

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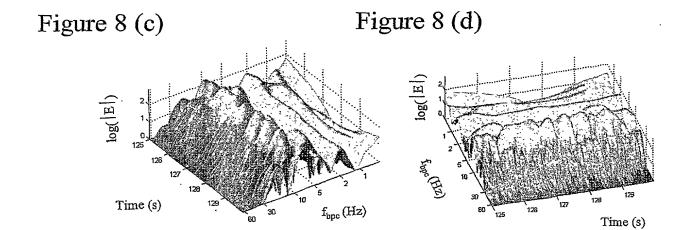
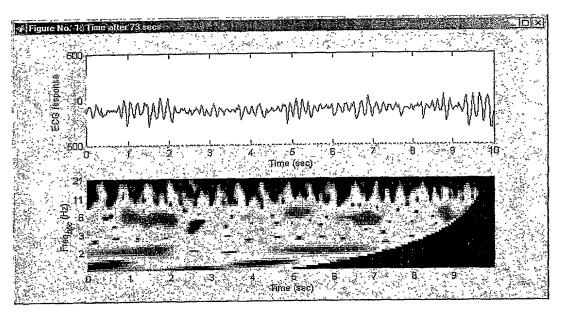
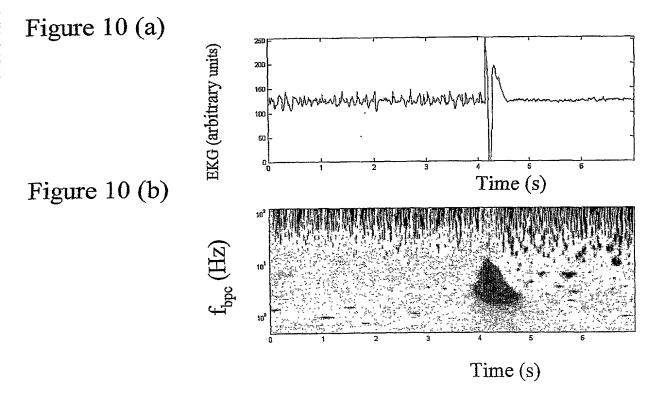


Figure 9





ROSC weights

Lower Frequ.

Figure 11 (a)

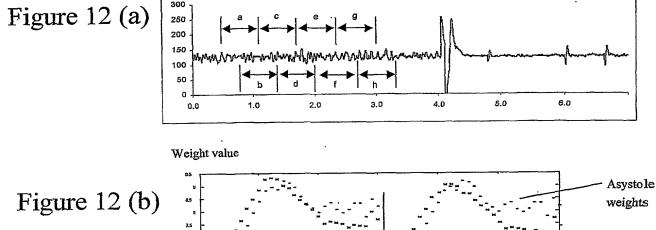
Figure 11 (b)

Separate 11 (b)

Figure 11 (c)

Figure 11 (d)

Figu



Lower Frequ.

Higher Frequ.

Higher Frequ.

mean energy (log)

10/14

Figure 13 (a)

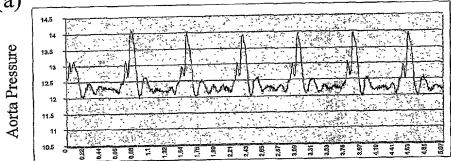


Figure 13 (b)

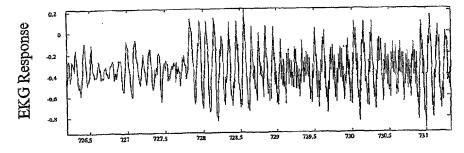


Figure 13 (c)

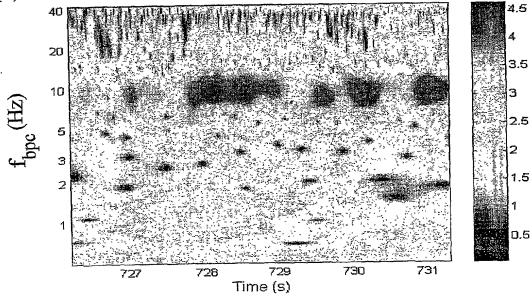


Figure 13 (d)

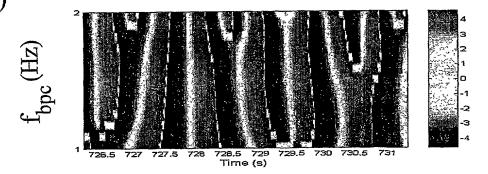


Figure 13 (e)

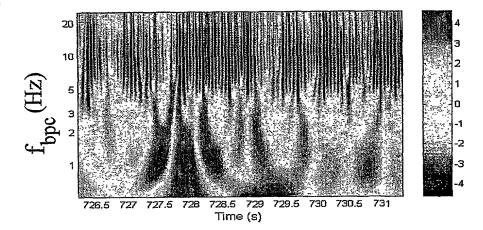
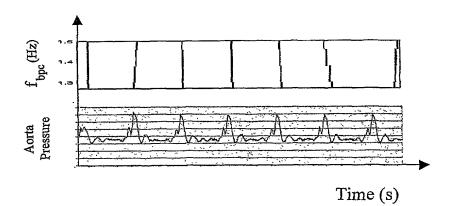
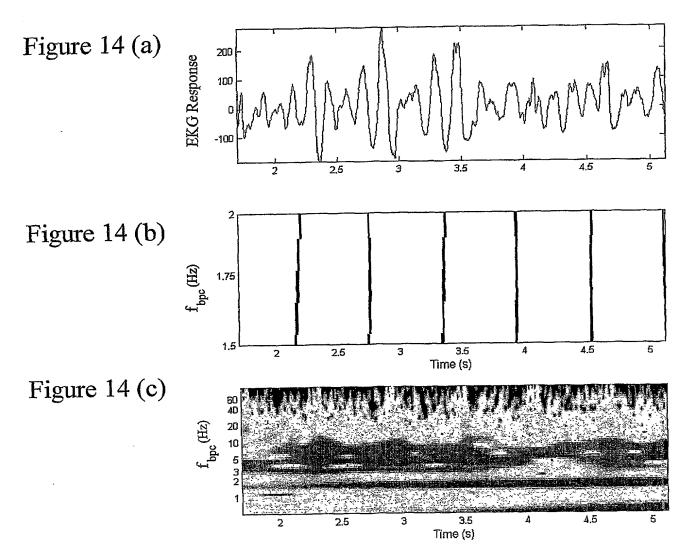


Figure 13 (f)



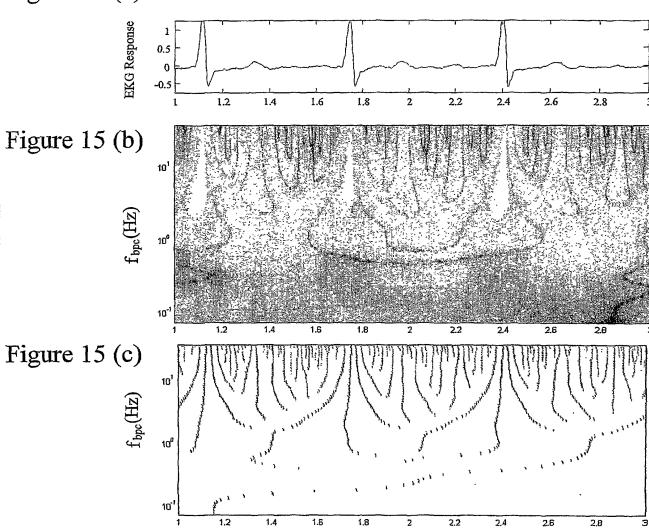
12/14



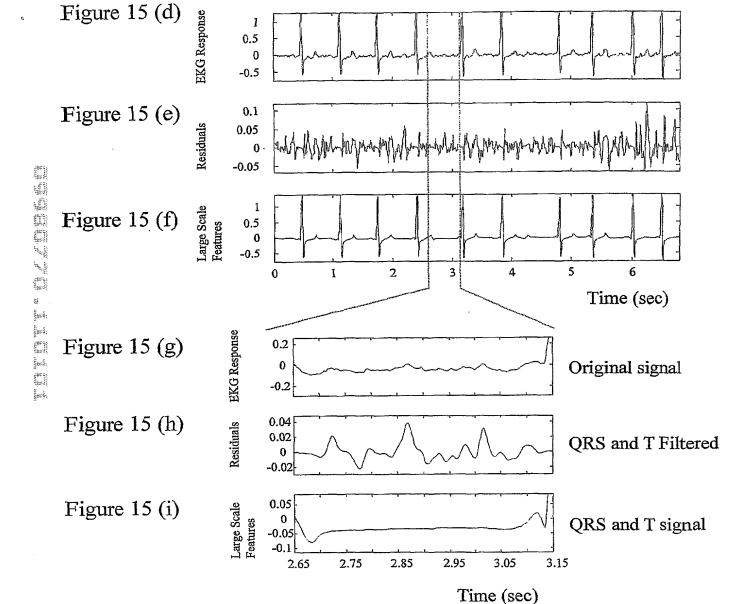
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Time (Seconds)

Figure 15 (a)



J.



NIXON PEABODY LLP No. Attorney's Docket

101 Federal Street Boston, Massachusetts 02110

Page 1 of 4

DECLARATION AND POWER OF ATTORNEY

As a below named inventor, I hereby declare that: My residence, post office address and citizenship are as stated below next to my name. I believe I am the original, first and sole inventor (if only one name is listed at 201) below or an original, first and joint inventor (if plural names are listed at 201-208 below) of the subject matter which is claimed and for which a patent is sought on the invention entitled:

	"Method of Analysis of Medical Signals"
which is d	escribed and claimed in:
	the specification attached hereto.
28	the specification in PCT Application Senal Number PCT/GB00/01675 / filed on 2 May 2000 /

I hereby state that I have reviewed and understand the contents of the above identified specification, including the claims, as amended by any amendment referred to above. I acknowledge the duty to disclose information which is material to the examination of this application in accordance with Title 37, Code of Federal Regulations, §1.56(a). I hereby claim foreign priority benefits under Title 35, United States Code, §119 of any foreign application(s) for patent or inventor's certificate listed below and have also identified below any foreign application for patent or inventor's certificate having a filing date before that of the application on which priority is claimed.

Prior Foreign/PCT Applications and Any Priority Claims Under 35 U.S.C. §119:						
Application No.	Filing Date	Country	Priority Claimed under 35 U.S.C. §1197			
9910019.0 ~	1 May 1999 -	United Kingdom	PAYES DINO			
9916499.8 ~	15 July 1999	United Kingdom >	MYES DNO			
9919677.6	20 August 1999 .	United Kingdom	MYES DNO			
9923110.2	1 October 1999 -	United Kingdom	BYES UNO			
0003711.9 🔛	17 February 2000	United Kingdom /	MYES ONO			

I hereby claim the benefit under 35 U.S.C. §120 of any United States application(s) or PCT international application(s) designating the United States of America that is/are listed below, and insofar as the subject matter of each of the claims of this application is not disclosed in that/those prior application(s) in the manner provided by the first paragraph of 36 U.S.C. §112, I acknowledge the duty to disclose material information as defined in 37 CFR §1.56(a) which occurred between the thing date of the prior application(s) and the national or PCT international filing date of this application.

Prior U.S.	Applications (or PCT International Ap under 35 U		signating the	U.S-Benetit
	U.S. Ap	plications	s	tatus (Check	One)
Application !	Serial No.	U.S. Filing Date	Patented	Pending	Abandoned

PCT App	lications Des	ignating the U.S.			
Application No.	Filing Date	U.S. Serial No. Assigned			

CLAIM FOR BENEFIT OF PRIOR U.S. PROVISIONAL APPLICATION(S) (35 U.S.C. §119(e))

I hereby claim the benefit under Title 35, United States Code, §119(e) of any United States provisional application(s) listed below:

Applicant	Provisional Application Number	Filing Date

POWER OF ATTORNEY: As a named inventor, I hereby appoint the following attorney(s) with full powers of association, substitution and revocation to prosecute this application and transact all business in the Patent and Tradernark Office connected therewith.

Romeid L.

Ronald L Eisenstein Nicola L.M. Valor William T French (Reg No 30,628) (Reg No 17 (50) (Reg No 16,297) David S. Rasmok. Georgia Evenia Gunnar G. Leinberg (Reg No 34 235) (Reg No 44,957)

Michael L Goldman (Reg No 30,727) Lisa A Dolak (Reg No 35,491) Edwin V Merke) (Key No 40,087)

Jaseph Noto

(Reg. No. 32,:03)

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NIXON PEABODY-LLP

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David S. Resnick (617) 345-6057

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	FULL NAME OF	LAST NAME	FIRST NAME	MIDDLE NAME
	INVENTOR	WATSON	JAMES	NICHOLAS
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			UNITED KINGDOM	A BRITISH // SUBJECT
	POST OFFICE ADDRESS	POST OFFICE ADDRESS	CITY	STATE OR COUNTRY AND ZIP CODE
		34 FOWLER TERRACE	EDINBURGH CRX	UNITED KINGDOM EH11 1DA

203	FULL NAME OF INVENTOR	LAST NAME	FIRST NAME	MIDDLE NAME
	RESIDENCE & CITIZENSHIP	CITY	STATE OR FOREIGN COUNTRY	COUNTRY OF CITIZENSHIP
	POST OFFICE ADDRESS	POST OFFICE ADDRESS	CITY	STATE OR COUNTRY AND ZIP CODE

I hereby further declare that all statements made herein of my own knowledge are true and that all statements made on information and belief are believed to be true; and further, that these statements were made with the knowledge

that wilful false statements and the like so made are punishable by fine or imprisonment, or both, under Section 1001 of Title 18 of the United States Code, and that such wilful false statements may jeopardize the validity of the application or any patent issued thereon.

Signature of Inventor 201	Date: 30 OCT 2001
Signature of Inventor 202	Date: 30 - 10 - 200 \
Signature of Inventor 203	Date: